Testing for mildly versus strongly misspecified models

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Checklist

1. For all authors...
   (a) Do the main claims made in the abstract and introduction accurately reflect the paper's contributions and scope? [Yes]
   (b) Did you describe the limitations of your work? [Yes]
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2. If you are including theoretical results...
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3. If you ran experiments...
   (a) Did you include the code, data, and instructions needed to reproduce the main experimental results (either in the supplemental material or as a URL)? [Yes] Included as a link to github
   (b) Did you specify all the training details (e.g., data splits, hyperparameters, how they were chosen)? [Yes] Everything is specified (and all details are given in the code on github)
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   (d) Did you include the total amount of compute and the type of resources used (e.g., type of GPUs, internal cluster, or cloud provider)? [N/A] The time to compute is of order of seconds.
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5. If you used crowdsourcing or conducted research with human subjects...
   (a) Did you include the full text of instructions given to participants and screenshots, if applicable? [N/A]

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(b) Did you describe any potential participant risks, with links to Institutional Review Board (IRB) approvals, if applicable? [N/A]

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A Derivation of D and V for multinomial model

We consider the following probability model. Given counts of words $n_k^{(\ell)}$, $\sum_{k=1}^p n_k^{(\ell)} = n^{(\ell)}$ for separate (independent) texts $\ell = 1, \ldots, N$ by the same author, assume that the corresponding probabilities are the same in all texts given by θ_k, $\sum_{k=1}^p \theta_k = 1$. So here we have $N = 11$ books, with $n^{(\ell)}$ words in each book. The multinomial model for such that has the following likelihood:

$$L(\theta; (n_k^{(\ell)})) = \prod_{\ell=1}^{N} \prod_{k=1}^{p} \theta_k^{n_k^{(\ell)}}.$$

Only $p-1$ unknown parameters are independent. Then,

$$\ell(\theta) = \sum_{\ell=1}^{N} \sum_{k=1}^{p-1} n_k^{(\ell)} \log \theta_k + n_p^{(\ell)} \log \left(1 - \sum_{k=1}^{p-1} \theta_k\right)$$

and for $j = 1, 2, \ldots, p-1$, denoting $N_j = \sum_{\ell=1}^{N} n_j^{(\ell)}$:

$$\begin{aligned}
\frac{\partial \ell(\theta)}{\partial \theta_j} &= N_j/\theta_j - N_p/\left(1 - \sum_{k=1}^{p-1} \theta_k\right) \\
\frac{\partial^2 \ell(\theta)}{\partial \theta_j^2} &= -N_j/\theta_j^2 - N_p/\left(1 - \sum_{k=1}^{p-1} \theta_k\right)^2 \\
\frac{\partial^2 \ell(\theta)}{\partial \theta_j \partial \theta_m} &= -N_p/\left(1 - \sum_{k=1}^{p-1} \theta_k\right)^2
\end{aligned}$$

so, using $Cov(n_j^{(\ell)}, n_k^{(\ell)}) = -n^{(\ell)} \theta_j \theta_k$, we have

$$\begin{aligned}
V_{j,m}(\theta) &= E\left(\frac{\partial \ell(\theta)}{\partial \theta_j} \frac{\partial \ell(\theta)}{\partial \theta_m}\right) \\
&= \mathbb{E}\left(N_j/\theta_j - N_p/\left(1 - \sum_{k=1}^{p-1} \theta_k\right)\right)\left(N_m/\theta_m - N_p/\left(1 - \sum_{k=1}^{p-1} \theta_k\right)\right) \\
&= \sum_{\ell=1}^{N} [Cov(n_j^{(\ell)}/\theta_j, n_m^{(\ell)}/\theta_m) + Var(n_p^{(\ell)}/\theta_p) - Cov(n_p^{(\ell)}/\theta_p, n_j^{(\ell)}/\theta_j) - Cov(n_p^{(\ell)}/\theta_p, n_m^{(\ell)}/\theta_m)]
\end{aligned}$$

with $\theta_p = 1 - \sum_{k=1}^{p-1} \theta_k$, and on the diagonal

$$\begin{aligned}
V_{j,j}(\theta) &= E\left(\frac{\partial \ell(\theta)}{\partial \theta_j}\right)^2 \\
&= \sum_{\ell=1}^{N} \mathbb{E}\left(n_j^{(\ell)}/\theta_j - n_p^{(\ell)}/\theta_p\right)^2 \\
&= \sum_{\ell=1}^{N} \left[Var(n_j^{(\ell)})/\theta_j^2 + Var(n_p^{(\ell)})/\theta_p^2 + -2Cov(n_j^{(\ell)}, n_p^{(\ell)})/[\theta_j \theta_p]\right].
\end{aligned}$$

Also,

$$\begin{aligned}
D_{jj}(\theta) &= \mathbb{E}N_j/\theta_j^2 + \mathbb{E}N_p/\theta_p^2 = \sum_{\ell=1}^{N} n^{(\ell)}[1/\theta_j + 1/\theta_p], \\
D_{jm}(\theta) &= \mathbb{E}N_p/\theta_p^2 = \sum_{\ell=1}^{N} n^{(\ell)}/\theta_p].
\end{aligned}$$

The (p)MLE is

$$\hat{\theta}_j = N_j / [\sum_{\ell=1}^{N} n^{(\ell)}], \quad j = 1, \ldots, p.$$

Denote $X_j^{(\ell)} = n_j^{(\ell)} / \hat{\theta}_j - n^{(\ell)}$, then

$$\begin{aligned} \hat{V}_{j,m} &= V_{y,jm}(\hat{\theta}) = \sum_{\ell=1}^{N} \left[X_j^{(\ell)} X_m^{(\ell)} + [X_p^{(\ell)}]^2 + -X_j^{(\ell)} X_p^{(\ell)} - X_m^{(\ell)} X_p^{(\ell)} \right], \\ \hat{V}_{j,j} &= V_{y,jj}(\hat{\theta}) = \sum_{\ell=1}^{N} \left[[X_j^{(\ell)}]^2 + [X_p^{(\ell)}]^2 + -2X_j^{(\ell)} X_p^{(\ell)} \right] \end{aligned}$$

B Statistical analysis of texts in R

The code used to analyse simulated and text data, with all preprocessing details and the list of books by A. Conan Doyle used in the analysis, is available on `https://github.com/nataliabochkina/TestMisspecifiedModel`.